

A generic model architecture for perceptual grouping

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Jonas Kubilius^{1,2}, Johan Wagemans², & Hans P. Op de Beeck¹ ¹Laboratory of Biological Psychology, KU Leuven (Belgium)

²Laboratory of Experimental Psychology, KU Leuven (Belgium)

Goal and assumptions

Image segmentation (figure-ground assignment) using perceptual organization

Converging lines of research

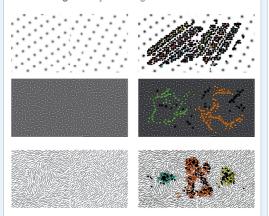
- Psychophysics and neuropsychology of spatial form defined by luminance, color, texture, motion, and binocular disparity (Regan, 2000)
- Neurophysiology on figure-ground assignment (cf. Self & Roelfsema, 2013; Kogo & van Ee, 2013)
- Perceptual organization model by Geisler & Super
- Graph-based segmentation, esp. normalized cuts (Shi & Malik, 2000)

Assumptions

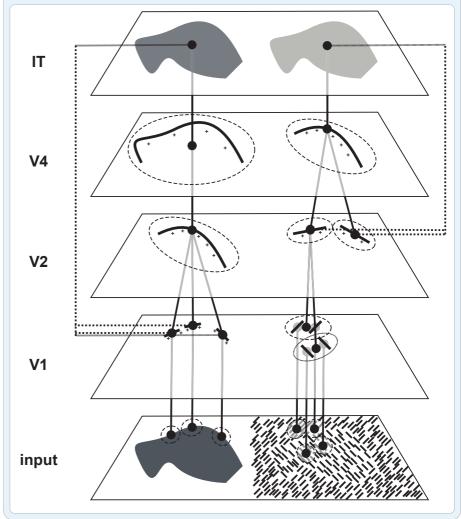
- Uniform acuity (central vision)
- Pre-attentive
- No color
- No motion
- No saccades

Partial implementation

- 1. Multiscale edge detection
- 2. Initial location sampling: 1% maximally responding locations (3 iterations)
- 3. Similarity (grouping strength) computation (rules: proximity, good continuation, orientation and contrast similarity; 3 iterations)
- 4. Final location sampling: 1000 maximally responding
- 5. Clustering into objects using max threshold



gmin: A hierarchy of detecting discontinuities and grouping



Open questions

- Best way to extract features? Even edge detection is hard.
- What is the order of computations? In particular, does figure-ground assignment start before contour integration is finished?
- What is the principled way to integrate different cues?
- Excitation / inhibition: What is the grouping stopping criterion?

Current limitations

- Number of objects predefined Equivalent to setting a grouping threshold. BUT: representation is hierarchical, so threshold varies depending on the task
- Grouping principles chosen per image BUT: grouping might be task-dependent (e.g., do people really see contour integration stimuli without any topdown knowledge?)
- Location sampling density alters performance Might be alleviated by Sharon et al. (2006)
- Clustering performance not stable Need a better method for clustering
- Does not work on real images Largely due to undersampling

A note on testing models

When a segmentation model makes a mistake, is it due to:

- lack of top-down knowledge, or
- poor bottom-up implementation?
- To avoid such confounds, top-down-free artificial scenes might be useful





"indoor" scene



